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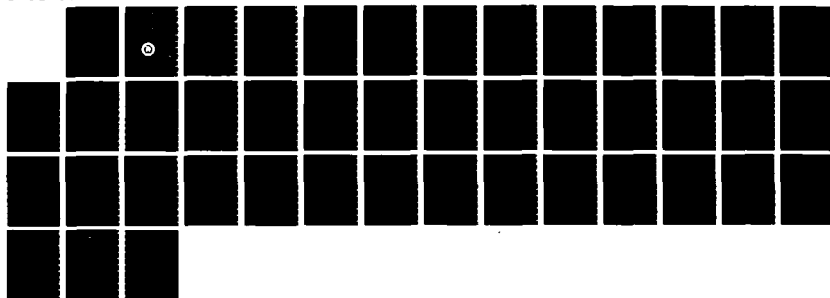
AN ALGORITHM FOR THE COMPUTATION OF GENERALIZED  
LIKELIHOOD OR SELF-CRITIC. (U) RENSSELAER POLYTECHNIC  
INST TROY NY SCHOOL OF MANAGEMENT T A DELAWANTY ET AL.  
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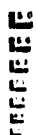
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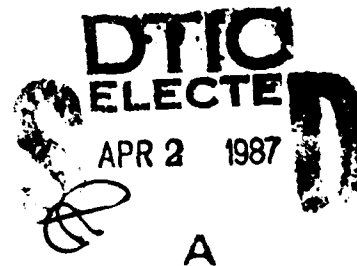
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**AN ALGORITHM FOR THE COMPUTATION OF GENERALIZED  
LIKELIHOOD OR SELF-CRITICAL ESTIMATORS  
FOR BINARY DATA**

by

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**An Algorithm for the Computation of Generalized Likelihood  
or Self-Critical Estimators for Binary Data**

by

**T.A. Delehanty\***

and

**A.S. Paulson\***

**Rensselaer Polytechnic Institute**



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## Documentation for Self-Critical Binary Program

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### 1. Purpose

Subroutine BINARY is an implementation of the self-critical estimation procedure of Paulson, Presser and Lawrence (1983). The logistic, Gaussian or Type I extreme value distribution may be selected as tolerance distribution. Estimates are expressed in location-scale form on entry and exit, but results may be printed out in regression form, location-scale form, or in both forms (see Section 3 for a description of the two parametrizations). It is possible to hold location parameters constant during the estimation procedure. Estimation is accomplished by a Newton-Raphson method.

### 2. Specification (FORTRAN)

SUBROUTINE BINARY (N, IX, X, IA, NPAR, ISUB, ISTART, IDEP, IDIST, C, RELTOL, ABSTOL, MAXIT, IPRINT, IFLAG, BETA, KLOGL, ICOV, COV, LMEN, MEMORY, IFAULT)

### 3. Description

The routine is applicable to the following modeling situation: For  $i = 1, \dots, n$  ( $n$  is the sample size), let  $v_i$  be the stress variable,  $a_i$  a zero-one indicator of withstand or failure, and  $X_i = (x_{i1}, \dots, x_{ip})^T$  a column vector of covariates. A constant is incorporated as a covariate identically equal to unity. The case  $p=0$  is possible, but should be rare.

The tolerance distribution and density,  $F(v_i)$  and  $f(v_i)$  respectively, depend on the covariates, a scale parameter  $\sigma$ , and a vector of location parameters  $\beta = (\beta_1, \dots, \beta_p)^T$  as follows, where

$$u_1 = \frac{(v_1 - \beta' X_1)}{\sigma},$$

- logistic,  $F(v_1) = \frac{\exp(u_1)}{(1 + \exp(u_1))}$ ,

$$f(v_1) = \frac{1}{|\sigma|} \frac{\exp(u_1)}{(1 + \exp(u_1))^2};$$

- Gaussian,  $F(v_1) = \Phi(u_1)$

( $\Phi$  is the standard Normal distribution function),

$$f(v_1) = \frac{1}{|\sigma| \sqrt{2\pi}} \exp(-\frac{1}{2} u_1^2);$$

- extreme value,  $F(v_1) = 1 - \exp(-\exp(u_1))$ ,

$$f(v_1) = \frac{1}{|\sigma|} \exp(u_1 - \exp(u_1)).$$

In principle, it is possible for  $\sigma$  to be negative, but this contradicts the physical notion of a stress variable.

In the method of maximum likelihood, parameters  $\theta$  are estimated by maximizing the log likelihood,

$$L(\theta) = \sum_{i=1}^n [a_i \log F(v_i) + (1-a_i) \log S(v_i)],$$

where  $S = 1-F$ . The self-critical procedure depends on a user-specified quantity  $c$ , and estimates  $\theta$  by solving the system

$$\sum_{i=1}^n f^c(v_i) \left[ a_i \frac{\partial}{\partial \theta} \log F(v_i) + (1-a_i) \frac{\partial}{\partial \theta} \log S(v_i) \right] = 0.$$

When  $c = 0$ , maximum likelihood estimates are obtained. As  $c$  increases from zero, increasingly robust estimators result. The sensitivity of parameter estimates to departures from model assumptions can be examined by starting with  $c = 0$  and refitting the model for increasing values of  $c$ . Negative values of  $c$  seem less useful.



#### 4. Numerical Method

For the purpose of estimation, the routine (internally) reparameterizes the problem in a "regression" form. In this parameterization,  $F(\cdot)$  depends on

$$w_1 = \alpha v_1 + \mu' X_1$$

instead of

$$u_1 = \frac{v_1 - \beta' X_1}{\sigma}.$$

The reparameterization offers two advantages:

- 1) Computation of the necessary partial derivatives is simple;
- 2) For maximum likelihood estimation with a logistic tolerance distribution, it can be shown that the regression parameterization results in a concave maximization problem. We anticipate that it will have fairly good properties in the more complicated cases.

It has a disadvantage in terms of potential ill-conditioning of the Hessian (or Jacobian) matrix, so that the input data should be sensibly scaled (see Section 11). Parameters are assumed to be expressed in the easier to understand location-scale form on entry, and are transformed back to location-scale form on exit, so the user need not worry about details of reparameterization.

Let

$$S_{\theta 1} = f^C(v_1) \left[ a_1 \frac{\partial}{\partial \theta} \log F(v_1) + (1 - a_1) \frac{\partial}{\partial \theta} \log S(v_1) \right].$$

The gradient used in Newton-Raphson iteration has  $\theta^h$  component

$$g_{\theta} = n^{-1} \sum_{i=1}^n S_{\theta 1}, \text{ and the Hessian has } (\theta, \theta') \text{ component } H_{\theta\theta'} = n^{-1} \sum_{i=1}^n \frac{\partial}{\partial \theta}, S_{\theta 1}.$$

(the scaling by  $n^{-1}$  should be helpful when  $n$  is large).

When  $c = 0$ , the estimated asymptotic covariance matrix of the estimators is  $n^{-1}(-H)^{-1}$ , while when  $c \neq 0$  it is  $n^{-1}H^{-1}VH^{-1}$ , where  $V$  has  $(\theta, \theta')$  component

$V_{\theta\theta'} = n^{-1} \sum_{i=1}^n S_{\theta i} S_{\theta' i}$ . The estimated asymptotic covariance matrix is ex-

pressed in location-scale form by pre- and post-multiplying it by the Jacobian of the underlying parameter transformation.

Newton-Raphson iteration converges when

$$\max_1 \left| \theta^{(k)} - \theta_1^{(k-1)} \right| < \text{ABSTOL}$$

and

$$\max_1 \left| \frac{\theta_1^{(k)}}{\theta^{(k-1)}} \right| < \text{RELTOL},$$

where superscripts represent iteration numbers and ABSTOL and RELTOL are user-supplied. The user also specifies a maximum allowable number MAXIT of iterations.

Solution of linear equations and matrix inversion is accomplished by subroutines DECOMP and SOLVE, taken from Forsythe, Malcolm and Moler (1977). These routines are of high numerical quality, and provide an estimate of the condition number of the input matrix, which is often useful. In the interest of portability, iterative improvement is not used.

## 5. Parameters

### 5.1 Input Parameters

- N - INTEGER  
Sample size. Unchanged on exit.
- IX - INTEGER  
Row dimension of data matrix X. Unchanged on exit.

**X -** REAL array of DIMENSION (IX, q), where  $q \geq N$ . Data matrix containing the dependent (stress) variable and all covariates. Each column corresponds to one observation. If parameters are held fixed, values of the dependent variable will be changed, but their input values restored. Unchanged on exit.

**IA -** INTEGER array of DIMENSION N. Withstand or failure is indicated for observation I according as  $IA(I) = 0$  or  $IA(I) \neq 0$ . Unchanged on exit.

**NPAR -** INTEGER. Number of parameters in the model. Unchanged on exit.

**ISUB -** INTEGER array of DIMENSION (NPAR). Indicates rows of data matrix X corresponding to parameters (the scale parameter is taken to correspond to the dependent variable). For instance, if  $ISUB(3) = 5$ , the third parameter corresponds to the fifth row of X. Unchanged on exit.

**ISTART -** INTEGER array of DIMENSION (NPAR). Indicates the status of parameters in the model, and whether a starting value is to be supplied. If  $ISTART(I)$  is equal to  
 0, BETA(I) is to be estimated, and its input value is to be disregarded;  
 1, BETA(I) is to be estimated, and its input value is to be used as starting value;  
 2, BETA(I) is to be held constant at its input value.  
 (See Section 11.2 for comments on starting values; the scale of parameter is not allowed to be held constant.)  
 Unchanged on exit.

**IDEP -** INTEGER. Row of dependent (stress) variable in X. Unchanged on exit.

**IDIST -** INTEGER. Indicates the tolerance distribution desired. If IDIST equals  
 1, logistic distribution will be used;  
 2, Gaussian distribution will be used;  
 3, extreme value distribution will be used.  
 Unchanged on exit.

**C -** DOUBLE PRECISION User-supplied constant for self-critical estimation. Unchanged on exit.

**RELTOL** - **DOUBLE PRECISION.**  
 Relative convergence tolerance for Newton-Raphson iteration  
 (see Section 4).  
 Unchanged on exit.

**MAXIT** - **INTEGER.**  
 Maximum allowable number of Newton-Raphson iteration.  
 Unchanged on exit.

**IPRINT** - **INTEGER.**  
 Output unit number. If  $IPRINT \leq 0$ , no output will be produced.  
 If  $IPRINT > 0$ , standard output will be produced on logical  
 output unit IPRINT.  
 Unchanged on exit.

**IFLAG** - **INTEGER.**  
 Indicator for form of output.  
 If  $IFLAG < 0$ , the standard output summary will be in regression  
 form. If  $IFLAG = 0$ , output summaries will be produced for  
 both regression and location-scale parameterizations. If  
 $IFLAG > 0$ , the standard output summary will be in location-  
 scale form. (See Section 11.3).  
 Unchanged on exit.

## 5.2 Input/Output (and associated dimension) Parameters.

**BETA** - **DOUBLE PRECISION** array of **DIMENSION (NPAR).**  
 On entry, contains starting values as specified by ISTART.  
 On exit, contains parameter estimates, in location scale form.

**XLOGL** - **DOUBLE PRECISION.**  
 On exit, contains the log likelihood if  $c = 0$  (maximum  
 likelihood estimation), and zero otherwise.

**ICOV** - **INTEGER.**  
 Row dimension for COV.  
 Unchanged on exit.

**COV** - **DOUBLE PRECISION** array of **DIMENSION (ICOV, q),** where  $q \geq NPAR$ .  
 On exit, estimated asymptotic covariance information for the  
 parameters. The diagonal contains standard errors, the strict  
 lower triangle correlations, and the strict upper triangle  
 covariances. If a parameter is held constant, all correspond-  
 ing entries are zero. The covariance matrix will be for the  
 location-scale parameterization rulers  $IFLAG < 0$ , when it will  
 be in regression form. See Section 11.3.

### 5.3 Workspace (and associated dimension) parameters

**LMEM** - INTEGER.

Length of work array MEMORY, as declared in calling program unit.  
 $LMEM \geq NPAR + 4*NACT*(1+NACT) + 2*max(NFIX, 2*NACT)$ , where  
NACT is the number of parameters estimated and NFIX = NPAR-NACT  
is the number of parameters held constant.  
Unchanged on exit.

**MEMORY** - INTEGER array of DIMENSION (LMEM).  
Used as workspace.

### 5.4 Diagnostic Parameter

**IFAU**LT - INTEGER.

Unless the routine finds an error or gives a warning, IFAULT  
will be 0 on exit. See Section 6.

## 6. Error Indications and Warnings

Errors or warnings specified by the routine:

**IFAU**LT < 0 IF IFAULT = -I, I > 0, then an arithmetic exception was about to occur on observation I, while computing partial derivatives. This failure should be rare. If the data have been sensibly scaled, and the starting values are not too bad (see Sections 11.1-11.2), the probable cause is a data error. The routine stops as soon as the error is detected, and output parameter values are not of interest.

**IFAU**LT = 1 Input parameter outside expected range. This failure will occur if, on entry,  $N < 1$ ,  $NPAR < 1$ ,  $IX < NPAR$ ,  $ICOV < NPAR$ ,  $RELTOL \leq 0$ ,  $ABSTOL < 0$ ,  $MAXIT < 1$ ,  $IDIST < 1$ ,  $IDIST > 3$ ,  $IPLAG \leq 0$  and  $IPRINT \leq 0$  (see Section 11.3),  $ISTART(I) < 0$  or  $ISTART(I) > 2$  for some I,  $ISTART(I) = 2$  for all I,  $ISUB(I) \neq IDEP$  for all I, or  $ISUB(I) = IDEP$  and  $ISTART(I) = 2$  for some I. (The last restriction means that the scale parameter cannot be held constant.) The routine stops without doing any calculations.

**IFAU**LT = 2 Insufficient workspace. This failure will occur if, on entry,  $LMEM < NPAR + 4*NACT*(1+NACT) + 2*MAX(NFIX, 2*NACT)$ , where NACT is the number of parameters estimated ( $ISTART(I) < 2$ ), and NFIX = NPAR - NACT is the number of parameters held constant ( $ISTART(I) = 2$ ). The routine stops without doing any calculations.

**IFAU**LT = 3 The Hessian matrix has become numerically singular during Newton-Raphson iteration. The routine transforms estimates to location-scale form, restores values of the dependent variable if parameters were held fixed, and stops. Output parameter values are not generally of interest.

IPFAULT = 4 The Newton-Raphson iteration did not converge to within RELTOL and ABSTOL in the specified MAXIT iterations. The routine transforms estimates to location-scale form, restores values of the dependent variable if parameters were held fixed, and stops. Output parameter values are not generally of interest.

## 7. Auxiliary Routines

SUBROUTINE PREPAR(DX, N, X, NPAR, BETA, ISUB, ISTART, NACT, IACT, IMAP, PAR, NFIX, IPIXED, IDEP, WORK)

Prepares for iteration by setting up some indexing and working arrays, subtracting the effects of fixed parameters from the dependent variable, and setting initial active parameter values in regression form.

SUBROUTINE NEWTON(C, MAXIT, RELTOL, ABSTOL, DIFFER, XLOGL, NACT, IACT, PAR, IPIVOT, HESS, HESFAC, N, DX, X, IA, IPRINT, LWORK, WORK, LPAULT)

Carries out the Newton-Raphson iteration.

SUBROUTINE VARIAB(C, XLOGL, REGRES, DIFFER, VROUT, NACT, IACT, IMAP, PAR, IPIVOT, HESS, HESFAC, ICOV, NPAR, COV, DET, N, DX, X, IA, WORK)

Computes the estimated asymptotic covariance matrix, transforms it to location-scale form if requested, and computes correlations and standard errors.

SUBROUTINE EXPOST(REGRES, IDEP, NPAR, BETA, ISTART, NACT, PAR, IMAP, N, DX, X, NFIX, IPIXED, WORK)

Sets the output vector BETA, transforms to location-scale form if requested. If location-scale form is requested and parameters were held fixed, restore initial values of dependent variable.

SUBROUTINE RESULT (C, REGRES, IPRINT, IDIST, N, IDEP, NACT, NFIX, XLOGL, NPAR, BETA, ISUB, ISTART, ICOV, COV, DET)

Produces a standard output summary on logical output unit IPRINT.

The following routines are specific to particular tolerance distributions, and are declared EXTERNAL in BINARY. The first three, prefixed D, are passed to NEWTON and VARIAB. The next three, prefixed V, are passed to VARIAB.

```

SUBROUTINE DLOGBN(C, OBJECT, NACT, IACT, PAR, GRAD, HESS, N, IX, X,
IA,, IFAULT)
SUBROUTINE DGAUBN ( _____ " _____)
SUBROUTINE DEXVBN ( _____ " _____)

```

Compute gradient and Hessian for logistic, Gaussian and extreme value models, resp.

```

SUBROUTINE VLOGBN(C, NACT, IACT, PAR, V, N, IX, X, IA)
SUBROUTINE VGAUBN(_____ " _____)
SUBROUTINE VEXUBN(_____ " _____)

```

Compute V factor of estimated asymptotic covariance matrix for logistic, Gaussian and extreme value models, resp.

The following procedures perform general numerical tasks. They have been included in the interest of portability, although equivalents exist on many computer systems.

```

SUBROUTINE DECOMP(NDIM, N, A, UL, COND, IPUT, WORK)
SUBROUTINE SOLVE(NDIM, N, A, B, X, IPUT)

```

DECOMP decomposes a matrix into LU factors and estimates its condition. SOLVE solves a linear system, using the results of DECOMP. These routines are in Forsythe, Malcolm and Moler (1977), but the present versions include an extra argument so that A and B need not be overwritten.

```

SUBROUTINE DMXMLT(A, IA, N1, B, IB, N2, C, IC, N3, WORK, LWORK,
IFLAG, IFAULT)

```

Double precision matrix multiplication -  
 $A(N1 \times N3) = B(N1 \times N2) * C(N2 \times N3)$ , where B is overwritten if IFLAG < 0 and c is overwritten if IFLAG > 0.

```

DOUBLE PRECISION FUNCTION ALNORM (X, UPPER)

```

Algorithm as 66 (Hill, 1973) to compute tail areas of the standard Normal curve.

```

DOUBLE PRECISION FUNCTION RMILLS(X)

```

Computes  $Z(x)/Q(x)$ , the reciprocal of Mills' ratio, where x is a standard Normal variate. This procedure is based on procedures by Hill (1973) and Adams (1969).

## 8. References

Adams, A.G. (1969). Algorithm 39. Areas under the normal curve. Computer J., 12, 197-198.

Forsythe, G.E. Malcolm, M.A. and Moler, C.B. (1977). Computer Methods for Mathematical Computations. Englewood Cliffs, New Jersey: Prentice-Hall.

Hill, I.D. (1973). Algorithm AS 66. The normal integral. Applied Statistics, 22, 424-427.

Paulson, A.S., Presser, M.A. and Lawrence, C.E. (1983). Self-critical, and robust, procedures for the analysis of binary data. Operations Research and Statistics Report No. 37-83-P1, Department of Operations Research and Statistics, Rensselaer Polytechnic Institute.

## 9. Storage

There are no internally declared arrays.

## 10. Precision, Machine Dependent Constants

The routine was developed on an IBM computers. The data matrix is single precision to conserve storage. To convert to single precision, take the following steps, with the exception noted below:

- 1) Change all DOUBLE PRECISION declarations to REAL;
- 2) Replace references to double precision FORTRAN library functions with single precision versions, e.g., EXP replaces DEXP;
- 3) Replace double precision constants by their single precision versions, e.g., 1.0 replaces 1.0D0.

Note: In routines PREPAR and EXPOST, the dependent variable is adjusted for the effect of fixed parameters. These adjustments must be calculated in double precision



to avoid a loss of significant digits in  $X$  which could affect subsequent computations.

Procedures DLOGBN, VLOGBN, DGAUBN, VGAUBN, DEXUBN, VEXUBN, DECOMP, ALNORM and RMILLS use machine-dependent constants whose values are set in DATA statements. These constants may have to be altered for some computers. They are pointed out by comments in the program units.

## 11. Further Comments

### 11.1 Scaling, conditioning of Hessian matrix

The user should be aware of potential problems of ill-conditioning. Because a regression-type parameterization is used, calculation of the Hessian matrix involves operations similar to the formation of  $X^T X$ , where  $X$  is the data matrix. Unfortunately, the resulting Hessian is often rather ill-conditioned. If IPRINT > 0, an estimate of the Hessian's condition number is printed out at each iteration. Roughly speaking, a condition number in excess of  $10^7$  is worrisome, although condition numbers in excess of  $10^{10}$  can be tolerated when working in double precision on an IBM 3081. The user can and should avoid potentially excessive ill-conditioning by scaling the data matrix  $X$  before calling the routine.

Although it is not known how to optimally scale a problem, it seems that a nearly ideal scaling will be achieved if nonconstant variables are transformed to the range  $[-1, 1]$ , centered at zero. However, such precise scaling is often tedious. It seems most important to "equilibrate" the data matrix so that all variables have roughly the same (moderate) magnitude. Such equilibration is often simply accomplished by dividing by suitable powers of 10, e.g., expressing voltages in megavolts instead of kilovolts. Centering the variables will further reduce the condition number. One often centers covariates anyway, so that the "intercept"

parameter has a clear interpretation.

### 11.2. Starting values

Since Newton-Raphson procedure is employed, starting values are important. If no starting values are supplied by the user ( $ISTART(I) = 0$  for all  $I$ ), the routine will use unity for the scale parameter and zero for all location parameters. These starting values will not be acceptable unless the dependent variable has been centered and scaled. It is recommended that the user employ the mean and standard deviation of the dependent variable as starting values for the "intercept" parameter (if any) and scale parameter, respectively. If the model is sequentially refit with different values of  $c$ , it is recommended that estimates from the most recent call be used as starting values for the next call.

### 11.3 Output Flags

Because the routine will generally be called sequentially with different values of  $c$ , it is most convenient to always express entry and exit parameter values in the same form. The location-scale form is used, because that form affords the clearest interpretation. An inconsistency arises if  $IPLAG < 0$ , for then the output estimates are in location-scale form, but the covariance matrix is in regression form. Thus, if the user wants to make separate use of the output parameters of BINARY, the routine should be called with  $IPLAG \geq 0$ .

It is recommended that the routine be called with  $IPRINT > 0$ . If  $IPLAG \leq 0$ , the requirement  $IPRINT > 0$  is enforced. (When  $IPLAG < 0$ , the requirement is in keeping with the inconsistency mentioned above.) The standard output summary should

be sufficient for most applications. For the maximum flexibility in output and interpretation of results, call BINARY with IPRINT > 0, IFLAG = 0.

## 12. Example

The following program illustrates the use of BINARY, and the standard output produced when IPRINT > 0 and IFLAG = 0.

```

C.....BINA0001
C (OPTIONALLY ROBUST) BINARY DATA ESTIMATION ROUTINE. BINA0002
C LOGISTIC, EXTREME VALUE, OR GAUSSIAN TOLERANCE DISTRIBUTION BINA0003
C CAN BE USED. PARAMETERS ON INPUT/OUTPUT ARE IN LOCATION- BINA0004
C SCALE FORM, BUT ESTIMATION IS CARRIED OUT USING A BINA0005
C REGRESSION-TYPE REPARAMETERIZATION. IF PRINTOUT IS DESIRED BINA0006
C (IPRINT GT 0), RESULTS CAN BE PRINTED OUT IN REGRESSION BINA0007
C FORM, LOCATION-SCALE FORM, OR BOTH. ARGUMENT IFLAG BINA0008
C SPECIFIES THE OUTPUT DESIRED. BINA0009
C FAILURE CODES - BINA0010
C IFALT = 1 - INPUT ERROR BINA0011
C IFALT = 2 - INSUFFICIENT WORKSPACE SUPPLIED BINA0012
C IFALT = 3 - HESSIAN MATRIX IS NUMERICALLY SINGULAR BINA0013
C IFALT = 4 - NO CONVERGENCE IN MAXIT ITERATIONS BINA0014
C IFALT = -1 - AN EXCEPTION WAS ABOUT TO OCCUR WHILE BINA0015
C PROCESSING THE I TH OBSERVATION BINA0016
C.....BINA0017
C SUBROUTINE BINARY(N, IX, X, IA, NPAR, ISUB, ISTART, IDEP, IDIST, BINA0018
C C, RELTOL, ABSTOL, MAXIT, IPRINT, IFLAG, BETA, XLOGL, BINA0019
C ICOV, COV, LMEM, MEMORY, IFAULT) BINA0020
C EXTERNAL PROCEDURES FOR PARTICULAR TOLERANCE DISTRIBUTIONS BINA0021
C EXTERNAL OLOGBN, DGAUBN, DEXVBN, VLOGBN, VGAUBN, VEXVBN BINA0022
C ARGUMENTS BINA0023
C INTEGER N, IX, IA(N), NPAR, ISUB(NPAR), ISTART(NPAR), IDEP, IDIST, BINA0024
C MAXIT, IPRINT, IFLAG, ICOV, LMEM, MEMORY(LMEM), IFAULT BINA0025
C REAL X(IX,N) BINA0026
C DOUBLE PRECISION C, RELTOL, ABSTOL, BETA(NPAR), XLOGL, BINA0027
C COV(ICOV,NPAR) BINA0028
C LOCAL SCALARS BINA0029
C LOGICAL REGRES BINA0030
C INTEGER NACT, NFIX, IND, IND1, MIACT, MIFIX, MIMAP, MIPVT, MPAR, BINA0031
C MHFESS, MHFAC, MWORK BINA0032
C DOUBLE PRECISION DET, ZERO BINA0033
C CONSTANTS BINA0034
C DATA ZERO /0.000/ BINA0035
C BINA0036
C CHECK FOR INPUT ERRORS BINA0037
C IFALT = 1 BINA0038
C IF (N LT 1 OR NPAR LT 1 OR IX LT NPAR OR ICOV LT BINA0039
C NPAR) RETURN BINA0040
C IF (RELTOL LE ZERO OR ABSTOL LE ZERO OR MAXIT LT 1) BINA0041
C RETURN BINA0042
C IF (IDIST LT 1 OR IDIST GT 3) RETURN BINA0043
C IF (IFLAG LE 0 AND IPRINT LE 0) RETURN BINA0044
C BINA0045
C MEMORY MANAGEMENT, POSSIBLE RELATED INPUT ERRORS. CHECK BINA0046
C ISUB(=), ISTART(=) VECTORS. COUNT ACTIVE PARAMETERS. BINA0047
C NOTE SCALE PARAMETER CANNOT BE FIXED. BINA0048
C NACT = 0 BINA0049
C IND1 = 0 BINA0050
C DO 10 I = 1, NPAR BINA0051
C IND = ISUB(I) BINA0052
C IF (IND LT 1 OR IND GT IX) RETURN BINA0053
C IF (IND EQ IDEP) IND1 = 1 BINA0054
C IND = ISTART(I) BINA0055
C IF (IND LT 0 OR IND GT 2) RETURN BINA0056
C IF (IND NE 2) NACT = NACT + 1 BINA0057
C BINA0058
C 10 CONTINUE BINA0059
C IF (NACT EQ 0 OR IND1 EQ 0) RETURN BINA0060
C IF (ISTART(IND1) EQ 2) RETURN

```

```

C      NFIX = NPAR - NACT
C      CHECK IF WORKSPACE SIZE IS ADEQUATE, ALLOCATE IT
C      IF(AULT = 2)
C        IND = 2 * NACT
C        IND1 = IND * NACT
C        IF (LMEM .LT. NPAR + 2*NACT + IND + 2*IND1 + MAXO(2*NFIX, 2*IND))
C          RETURN
C        MIACI = 1
C        MIFIX = MIACI + NACT
C        MIMAP = MIACI + NPAR
C        MIPVT = MIMAP + NACT
C        MPAR = MIPVT + NACT
C        MHESH = MPAR + IND
C        MHFAC = MHESH + IND1
C        MWORK = MHFAC + IND1
C
C      PREPARE FOR ESTIMATION - SET UP SUBSCRIPT ARRAYS AND
C      STARTING VALUES FOR ACTIVE PARAMETERS, SUBTRACT EFFECT OF
C      FIXED PARAMETERS FROM DEPENDENT VARIABLE
C      CALL PREPAR(IX, N, X, NPAR, BETA, ISUB, ISTART, NACT,
C      1 MEMORY(MIACI), MEMORY(MIMAP), MEMORY(MPAR), NFIX,
C      2 MEMORY(MIFIX), IOEP, MEMORY(MWORK))
C
C      NEWTON-RAPHSON ITERATION WITH EXTERNAL ROUTINE PASSED FOR
C      FIRST AND SECOND PARTIALS
C      IF (IDIST.EQ. 1) CALL NEWTON(C, MAXIT, RELTOL, ABSTOL, DLOGBN,
C      1 XLOGL, NACT, MEMORY(MIACI), MEMORY(MPAR), MEMORY(MIPVT),
C      2 MEMORY(MHESH), MEMORY(MHFAC), N, IX, X, IA, IPRINT, 2*NACT,
C      3 MEMORY(MWORK), IFAULT)
C      IF (IDIST.EQ. 2) CALL NEWTON(C, MAXIT, RELTOL, ABSTOL, DGAUBN,
C      1 XLOGL, NACT, MEMORY(MIACI), MEMORY(MPAR), MEMORY(MIPVT),
C      2 MEMORY(MHESH), MEMORY(MHFAC), N, IX, X, IA, IPRINT, 2*NACT,
C      3 MEMORY(MWORK), IFAULT)
C      IF (IDIST.EQ. 3) CALL NEWTON(C, MAXIT, RELTOL, ABSTOL, DEXVBN,
C      1 XLOGL, NACT, MEMORY(MIACI), MEMORY(MPAR), MEMORY(MIPVT),
C      2 MEMORY(MHESH), MEMORY(MHFAC), N, IX, X, IA, IPRINT, 2*NACT,
C      3 MEMORY(MWORK), IFAULT)
C
C      ERROR HANDLING IF NEWTON-RAPHSON PROCEDURE FAILS.
C      ON ERROR, PRINT OUT MESSAGE IF IPRINT.GT. O, SET UP
C      THE FINAL CALL TO EXPST()
C      IF (IFAULT.EQ. O) GO TO 30
C      REGES = .FALSE.
C      IF (IPRINT.LE. O) GO TO 50
C      IF (IFAULT.GT. O) GO TO 20
C      IND = -IFAULT
C      WRITE (IPRINT,60) IND
C      GO TO 50
C20 IFAULT = IFAULT + 1
C      IF (IFAULT.EQ. 3) WRITE (IPRINT,70)
C      IF (IFAULT.EQ. 4) WRITE (IPRINT,80) MAXIT
C      GO TO 50
C
C      COMPUTE APPROXIMATE ASYMPTOTIC COVARIANCE MATRIX.
C      REGRES IS A FLAG FOR WHETHER OR NOT TO SET UP OUTPUT IN
C      REGRESSION FORM. STATEMENT 30, THE FIRST BRANCH POINT,
C      SETS UP REGRES FOR FIRST OUTPUT (IF REGRESSION AND
C      LOCATION-SCALE BOTH REQUESTED, REGRESSION GOES FIRST).
C      STATEMENT 40, THE SECOND BRANCH POINT, IS FOR REPEAT
C      BINA0061
C      BINA0062
C      BINA0063
C      BINA0064
C      BINA0065
C      BINA0066
C      BINA0067
C      BINA0068
C      BINA0069
C      BINA0070
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C      BINA0115
C      BINA0116
C      BINA0117
C      BINA0118
C      BINA0119
C      BINA0120

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C      CALL (ALWAYS LOCATION-SCALE)
30 REGRES = IFLAG 'LE' O
40 IF (IDIST EQ. 1) CALL VARIAB(C, XLOGL, REGRES, DLOGBN, VLOGBN,
1   NACT, MEMORY(MIACT)), MEMORY(MIMAP), MEMORY(MPAR),
2   MEMORY(MIPVT)), MEMORY(MHESS), MEMORY(MHFAC), ICOV, NPAR, COV,
3   DET, N, IX, X, IA, MEMORY(MWORK))
1   IF (IDIST EQ. 2) CALL VARIAB(C, XLOGL, REGRES, DGAUBN, VGAUBN,
2   NACT, MEMORY(MIACT)), MEMORY(MIMAP), MEMORY(MPAR),
3   MEMORY(MIPVT)), MEMORY(MHESS), MEMORY(MHFAC), ICOV, NPAR, COV,
4   DET, N, IX, X, IA, MEMORY(MWORK))
1   IF (IDIST EQ. 3) CALL VARIAB(C, XLOGL, REGRES, DEXVBN, VEXVBN,
2   NACT, MEMORY(MIACT)), MEMORY(MIMAP), MEMORY(MPAR),
3   MEMORY(MIPVT)), MEMORY(MHESS), MEMORY(MHFAC), ICOV, NPAR, COV,
4   DET, N, IX, X, IA, MEMORY(MWORK))
C      IF NEWTON-RAPHSON FAILED, GOODBYE.
C      IF (IFAILT NE. O) RETURN
C      PRODUCE STANDARD OUTPUT IF REQUESTED.
C      IF (IPRINT GT. O) CALL RESULT(C, REGRES, IPRINT, IDIST, N, IDEP,
1   NACT, NFIX, XLOGL, NPAR, BETA, ISUB, ISTART, ICOV, COV, DET)
C      IF OUTPUT DESIRED IN BOTH REGRESSION AND LOCATION-SCALE
C      FORMS, LOOP BACK, ELSE EXIT
C      REGRES = NOT REGRES
C      IF ( NOT REGRES AND IFLAG EQ. O) GO TO 40
C      IF ONLY REGRESSION FORM OUTPUT WAS REQUESTED, AN EXTRA
C      CALL TO EXPOST() IS NEEDED TO RESTORE LOCATION-SCALE FORM
C      IF (IFLAG LT. O) CALL EXPOST(REGRES, IDEP, NPAR, BETA, ISTART,
1   NACT, MEMORY(MPAR), MEMORY(MIMAP), N, IX, X, NFIX,
2   MEMORY(MIFIX)), MEMORY(MWORK))
C      60 FORMAT ('FAILURE - EXCEPTION ON OBSERVATION NO.', I5)
C      70 FORMAT ('FAILURE - HESSIAN MATRIX IS NUMERICALLY SINGULAR')
C      80 FORMAT ('WARNING - NO CONVERGENCE IN', I4, ' ITERATIONS - POSSIBL
1   FAILURE')
C      RETURN
C      END
C      PREPARE FOR ANALYSIS BY SETTING UP THE FOLLOWING ARRAYS -
C      IACT(*) - X(*) ROWS FOR ACTIVE PARAMETERS
C      IFIXED(*) - X(*) ROWS FOR FIXED PARAMETERS
C      IMAP(*) - POSITIONS IN BETA(*) OF ACTIVE PARAMETERS
C      PAR(*) - INITIAL VALUES OF ACTIVE PARAMETERS IN
C      REGRESSION FORM
C      ALSO SUBTRACT EFFECTS OF FIXED PARAMETERS FROM DEPENDENT
C      VARIABLE, USING WORK(*)
C      NOTE - DEPENDENT VARIABLE (SCALE) PARAMETER IS ALWAYS
C      PLACED FIRST IN PAR(*) FOR CONVENIENCE LATER
C      SUBROUTINE PREPAR(IX, N, X, NPAR, BETA, ISUB, ISTART, NACT, IACT,
1   IMAP, PAR, NFIX, IFIXED, IDEP, WORK)
C      ARGUMENTS

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PREP0075

INTEGER IX, N, NPAR, ISUB(NPAR), ISTART(NPAR), NACT, IACT(NACT),
1 IMAP(NACT), NFIX, IFIXED(I), IDEP
REAL X(IX,N)
DOUBLE PRECISION BETA(NPAR), PAR(NACT), WORK(I)
LOCAL SCALARS
INTEGER IND, IND1, LACT, ISCALE
DOUBLE PRECISION TEMP, ZERO, ONE
DATA ZERO, ONE /0.000, 1.000/

C
C
C      SET UP IACT(*), IFIXED(*), IMAP(*)
LACT = 0
IND = 0
ISCALE = 0
DO 20 I = 1, NPAR
  IF (ISTART(I) .EQ. 2) GO TO 10
  LACT = LACT + 1
  IMAP(LACT) = I
  IND = ISUB(I)
  IACT(LACT) = IND
  IF (IND1 .EQ. IDEP) ISCALE = I
  GO TO 20
10  IND = IND + 1
  IFIXED(IND) = ISUB(I)
  WORK(IND) = BETA(I)
20  CONTINUE

C
C      SWITCH PLACES SO SCALE IS FIRST ACTIVE PARAMETER
IND = IACT(1)
IACT(1) = IACT(ISCALE)
IACT(ISCALE) = IND
IND = IMAP(1)
IMAP(1) = IMAP(ISCALE)
IMAP(ISCALE) = IND

C
C      MAIN LOOP OVER SAMPLE IF PARAMETERS ARE FIXED, TO ADJUST
C      DEPENDENT VARIABLE (TO BE RESET ON EXIT FROM BINARY())
C      IF (NFIX .EQ. 0) GO TO 50
DO 40 I = 1, N

C
C      THE FOLLOWING INNER PRODUCT MUST BE ACCUMULATED IN DOUBLE
C      PRECISION...
TEMP = DBLE(X(IDEP,I))
DO 30 J = 1, NFIX
  IND = IFIXED(J)
  TEMP = TEMP - WORK(J) * X(IND,I)
30  CONTINUE
  X(IDEP,I) = SNGL(TEMP)

C
C      40 CONTINUE

C
C      SET STARTING VALUES
DO 60 I = 1, NPAR
60  PAR(I) = ZERO
  TEMP = ONE
  IND = IMAP(1)
  IF (ISTART(IND) .EQ. 1 AND BETA(IND) .NE. ZERO) TEMP = BETA(IND)
  PAR(1) = ONE / TEMP
  IF (NACT .EQ. 1) RETURN
  DO 70 I = 2, NACT
  IND = IMAP(I)

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      IF (ISTART(IND) EQ. 1) PAR(1) = -BETA(IND) / TEMP
      70 CONTINUE
C
      RETURN
      END
C.....
C      NEWTON-RAPHSON ITERATION
C      FAILURE CODES -
C      IF AULT = 1 - INSUFFICIENT WORKSPACE (LESS THAN 2 * NACT
C      LOCATIONS)
C      IF AULT = 2 - THE HESSIAN MATRIX IS NUMERICALLY SINGULAR
C      IF AULT = 3 - CONVERGENCE HAS NOT OCCURRED IN MAXIT ITNS
C      IF AULT = -1 - AN EXCEPTION WAS ABOUT TO OCCUR WHILE
C      PROCESSING THE 1 TH OBSERVATION IN DIFFER()
C.....
C      SUBROUTINE NEWTON(C, MAXIT, RELTOL, ABSTOL, DIFFER, XLOGL, NACT,
C      1 IACT, PAR, IPIVOT, HESS, HESFAC, N, IX, X, IA, IPRINT,
C      2 LWORK, WORK, IFAULT)
C      ARGUMENTS - DIFFER IS EXTERNAL ROUTINE FOR PARTIALS
C      INTEGER MAXIT, NACT, IACT(NACT), IPIVOT(NACT), N, IX, IA(N),
C      1 IPRINT, LWORK, IFAULT
C      REAL X(IX,N)
C      DOUBLE PRECISION C, RELTOL, ABSTOL, XLOGL, PAR(NACT),
C      1 HESS(NACT,NACT), HESFAC(NACT,NACT), WORK(LWORK)
C      LOCAL SCALARS
C      INTEGER ITER, IND
C      DOUBLE PRECISION GNORM, COND, ABSERR, RELERR, TEMP, XNEW, ZERO,
C      1 ONE
C      DATA ZERO, ONE /0.000, 1.000/
C
C      IFAULT = 1
C      IF (LWORK LT. 2*NACT) RETURN
C      IFAULT = 0
C      IND = NACT + 1
C      ITER = 0
C      IF (IPRINT GT. 0) WRITE (IPRINT,50) C, MAXIT, RELTOL, ABSTOL
C
C      LOOPING POINT FOR ITERATION - THE FIRST NACT LOCATIONS OF
C      WORK(*) ARE USED FOR GRADIENT/INCREMENT, WHILE THE NEXT
C      NACT LOCATIONS ARE WORKSPACE FOR DECOMP
C      10 ITER = ITER + 1
C      CALL DIFFER(C, XLOGL, NACT, IACT, PAR, WORK(1), HESS, N, IX, X,
C      1 IA, IFAULT)
C      IF (IFAULT LT. 0) RETURN
C
C      FACTOR THE HESSIAN INTO L * U, CHECK IF SINGULAR
C      CALL DECOMP(NACT, NACT, HESS, HESFAC, COND, IPIVOT, WORK(IND))
C      IF (COND + ONE NE. COND) GO TO 20
C      IFAULT = 2
C      RETURN
C
C      SOLVE THE SYSTEM HESS * INCREMENT = -GRADIENT,
C      OVERWRITING GRADIENT
C      20 GNORM = ZERO
C      DO 30 I = 1, NACT
C      TEMP = WORK(1)
C      WORK(1) = -TEMP
C      GNORM = GNORM + TEMP * TEMP
C      30 CONTINUE
C      GNORM = DSQRT(GNORM)

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C      CALL SOLVE(NACT, NACT, HESFAC, WORK, IPIVOT)
C      INCREMENT PARAMETERS, COMPUTE CONVERGENCE CRITERIA
      RELERR = ZERO
      ABSERR = ZERO
      DO 40 I = 1, NACT
        TEMP = WORK(I)
        ABSERR = DMAX1(ABSERR, DABS(TEMP))
        XNEW = PAR(I) + TEMP
        PAR(I) = XNEW
        IF (ONE + XNEW .NE. ONE) TEMP = TEMP / XNEW
        RELERR = DMAX1(RELERR, DABS(TEMP))
      40 CONTINUE
      IF (IPRINT .GT. 0) WRITE (IPRINT, 60) ITER, GNORM, RELERR, ABSERR,
      1 COND
C
C      CHECK CONVERGENCE, SET FAILURE CODE IF NONE
C      IF (RELERR .LT. RELTOL .AND. ABSERR .LT. ABSTOL) RETURN
      IF (ITER .LT. MAXIT) GO TO 10
      IF AULT = 3
C
      50 FORMAT ('1 C      MAX ITNS REL TOLER ABS TOLER'/D11.3, I9,
      1 2D11.2)
      60 FORMAT ('0 ITN GRAD NORM REL CHANGE ABS CHANGE'/I5, 3D12.2, /,
      1 1 ESTIMATED CONDITION NUMBER OF HESSIAN IS', D10.2)
      RETURN
      END
C.....
C      COMPUTATIONS FOR VARIABILITY OF THE ESTIMATORS
C      ESTIMATE ASYMPTOTIC COVARIANCE MATRIX, TRANSFORM IT FROM
C      REGRESSION TO LOCATION-SCALE FORM, COMPUTE ASYMPTOTIC
C      CORRELATIONS AND STD ERRORS
C.....
      SUBROUTINE VARIAB(C, XLDGL, REGRES, DIFFER, VROUT, NACT, IACT,
      1 IMAP, PAR, IPIVOT, HESS, HESFAC, ICOV, NPAR, COV, DET,
      2 N, IX, X, IA, WORK)
C      ARGUMENTS - DIFFER AND VROUT ARE SUBROUTINES
C      LOGICAL REGRES
      INTEGER NACT, IACT(NACT), IMAP(NACT), IPIVOT(NACT), ICOV, NPAR, N,
      1 IX, IA(N)
      REAL X(IX, N)
      DOUBLE PRECISION C, XLDGL, PAR(NACT), HESS(NACT, NACT),
      1 HESFAC(NACT, NACT), COV(ICOV, NPAR), DET, WORK(NACT)
C      LOCAL SCALARS
      LOGICAL FLAG
      INTEGER IND, IND1, IND2
      DOUBLE PRECISION ALPHA, ALPHA2, TEMP, TEMP1, ZERO, ONE, SMALL
C      MACHINE-DEPENDENT CONSTANT - SMALL SET SO THAT DEXP(X)
C      WILL CAUSE EXCEPTION IF X .LT. SMALL
      DATA ZERO, ONE, SMALL /0.000, 1.000, -180.000/
C
C      CALCULATE HESSIAN AT OPTIMAL POINT AND FACTOR IT
C      (THIS MAY BE WASTEFUL IN SOME CASES, BUT IT AVOIDS
C      SOME LOGICAL COMPLICATIONS.) WORK(*) USED AS SCRATCH
C      THE HESSIAN IS ASSUMED NONSINGULAR, AS IT SHOULD BE IF
C      THIS POINT IS REACHED FACTORIZATION TO HESFAC(*,*)
C      CALL DIFFER(C, XLDGL, NACT, IACT, PAR, WORK, HESS, N, IX, X, IA,
      1 IND)
C      CALL DECOMP(NACT, NACT, HESS, HESFAC, TEMP, IPIVOT, WORK)
C

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C      INVERT NEGATIVE OF HESSIAN, USING IPIVOT(*) AND
C      FACTORIZATION IN HESFAC(*,*). PLACE INVERSE IN HESS(*,*)
DO 20 J = 1, NACT
  DO 10 I = 1, NACT
    HESS(I,J) = ZERO
  HESS(J,J) = -ONE
  CALL SOLVE(NACT, NACT, HESFAC, HESS(I,J), HESS(I,J), IPIVOT)
20 CONTINUE
  IF (C EQ ZERO) GO TO 30

C      SECTION FOR ROBUST ANALYSIS -
C      PLACE PRODUCT (H INVERSE) * V * (H INVERSE) IN HESS(*,*)
C      FIRST COMPUTE V(*,*), PLACE IN COV(*,*)
C      CALL VRUT(C, NACT, IACT, PAR, COV, N, IX, X, IA)

C      MULTIPLY HESS * COV, OVERWRITING COV(*,*)
C      CALL DMXMT(COV, NACT, NACT, HESS, NACT, NACT, COV, NACT, NACT,
1      WORK, NACT, 1, IND)

C      MULTIPLY COV * HESS, OVERWRITING HESS(*,*)
C      CALL DMXMT(HESS, NACT, NACT, COV, NACT, NACT, HESS, NACT, NACT,
1      WORK, NACT, 1, IND)

C      ALL VALUES OF C - IF LOCATION-SCALE FORM, TRANSFORM
C      THE COVARIANCE MATRIX.
30 IF (REGRES) GO TO 80
  ALPHA = PAR(1)
  ALPHA2 = ALPHA * ALPHA
  FLAG = NACT EQ 1

C      LEFT-MULTIPLY HESS(*,*) BY JACOBIAN, RESULT TO HESFAC(*,*)
DO 50 J = 1, NACT
  TEMP = HESS(1,J) / ALPHA2
  HESFAC(1,J) = -TEMP
  IF (FLAG) GO TO 50
  DO 40 I = 2, NACT
    HESFAC(I,J) = PAR(I) * TEMP - HESS(I,J) / ALPHA
  40 CONTINUE

C      RIGHT-MULTIPLY HESFAC(*,*) BY TRANSPOSE OF JACOBIAN,
C      RESULT TO HESS(*,*)
DO 70 I = 1, NACT
  TEMP = HESFAC(1,I) / ALPHA2
  HESS(1,I) = -TEMP
  IF (FLAG) GO TO 70
  DO 60 J = 2, NACT
    HESS(I,J) = PAR(J) * TEMP - HESFAC(I,J) / ALPHA
  60 CONTINUE

C      BOTH PARAMETERIZATIONS - DIVIDE COVARIANCE MATRIX BY
C      SAMPLE SIZE
C      TEMP = DBLE(FLOAT(N))
DO 90 J = 1, NACT
  DO 90 I = 1, J
    HESS(I,J) = HESS(I,J) / TEMP
    HESS(J,I) = HESS(I,J)
90 CONTINUE

C      FIND DETERMINANT OF COVARIANCE MATRIX - FACTOR IT,
C      RESETING IPIVOT(*) AND HESFAC(*,*)

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      CALL DECOMP(NACT, NACT, HESS, HESFAC, TEMP, IPIVOT, WORK)
      DET = ZERO
      IF (TEMP + ONE EQ TEMP) GO TO 110
      IND = IPIVOT(NACT)
      DO 100 I = 1, NACT
        TEMP1 = HESFAC(I,1)
        IF (TEMP1 LT ZERO) IND = -IND
        DET = DET * DLOG(DABS(TEMP1))
      100 CONTINUE
C
C      TAKE THE NACT ROOT OF DETERMINANT. SET DET TO ZERO
C      IF IT UNDERFLOWS OR IS NEGATIVE.
C      DET = DET / DBLE(FLOAT(NACT))
      DET = DET / LT SMALL OR IND LT 0) DET = ZERO
      IF (DET GE SMALL AND IND GT 0) DET = DEXP(DET)
C
C      MOVE CONTENTS OF HESS(*,*) TO UPPER TRIANGLE OF COV(*,*)
      110 DO 120 J = 1, NPAR
        DO 120 I = 1, NPAR
          DO 120 COV(I,J) = ZERO
        DO 130 I = 1, NACT
          IND = IMAP(I)
          DO 130 J = 1, NACT
            IND1 = IMAP(J)
            IND2 = MINO(IND,IND1)
            IND1 = MAXO(IND,IND1)
            COV(IND2,IND1) = HESS(I,J)
          130 CONTINUE
C
C      PUT STD ERRORS ON DIAG OF COV(*,*). CORRELATIONS BELOW
      150 I = 1, NPAR
        TEMP = COV(I,1)
        IF (TEMP EQ ZERO) GO TO 150
        TEMP = DSORT(TEMP)
        COV(I,1) = TEMP
        IF (1 EQ 1) GO TO 150
        IND = 1 - 1
        DO 140 J = 1, IND
          TEMP1 = COV(J,J)
          IF (TEMP1 NE ZERO) COV(I,J) = COV(J,1) / (TEMP*TEMP1)
        140 CONTINUE
      150 CONTINUE
C
      RETURN
      END
C.....EXP00001
C      EX POST ADJUSTMENTS BEFORE EXIT
C      PUT OPTIMAL PARAMETERS IN BETA(*). IF LOCATION-SCALE
C      FORM, TRANSFORM THE PARAMETERS AND RESTORE DEPENDENT
C      VARIABLE TO INITIAL VALUES IF PARAMETERS WERE FIXED.
C.....EXP00006
C      SUBROUTINE EXPST(REGRES, IDEP, NPAR, BETA, ISTART, NACT, PAR,
C      1 IMAP, N, IX, X, NFIX, IFIXED, WORK)
C      ARGUMENTS
C      LOGICAL REGRES
C      INTEGER IDEP, NPAR, ISTART(NPAR), NACT, IMAP(NACT), N, IX, NFIX,
C      1 IFIXED(1)
C      REAL X(IX,N)
C      DOUBLE PRECISION BETA(NPAR), PAR(NACT), WORK(1)
C      LOCAL SCALARS

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 EXP00008  
 EXP00009  
 EXP00010  
 EXP00011  
 EXP00012  
 EXP00013  
 EXP00014  
 EXP00015

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EXP00016
EXP00017
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EXP00019
EXP00020
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RESU0001
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RESU0018
RESU0019

INTEGER IND
DOUBLE PRECISION TEMP, ONE
DATA ONE / 1.0D0/

C      INSERT WORKING PARAMETERS PAR(*) IN BETA(*), IN THE
C      SUITABLE FORM.
C      TEMP = PAR(1)
IND = IMAP(1)
IF (REGRES) BETA(IND) = TEMP
IF ( NOT REGRES) BETA(IND) = ONE / TEMP
IF (NACT.EQ.1) GO TO 20
DO 10 I = 2, NACT
  IND = IMAP(I)
  IF (REGRES) BETA(IND) = PAR(I)
  IF ( NOT REGRES) BETA(IND) = -PAR(I) / TEMP
10 CONTINUE

C      IF LOCATION-SCALE FORM AND PARAMETERS WERE HELD FIXED,
C      RESTORE DEPENDENT VARIABLE
C      20 IF (REGRES OR NFIX.EQ.O) RETURN
IND = O
DO 30 I = 1, NPAT
  IF (ISTART(I) NE 2) GO TO 30
  IND = IND + 1
  WORK(IND) = BETA(I)
30 CONTINUE
DO 50 I = 1, N

C      THE FOLLOWING INNER PRODUCT MUST BE ACCUMULATED IN DOUBLE
C      PRECISION
C      TEMP = DBLE(X(IDEP,I))
C      DO 40 J = 1, NFIX
C        IND = IFIXED(J)
C        TEMP = TEMP + WORK(J) * X(IND,I)
40 CONTINUE
X(IDEP,I) = SNGL(TEMP)

C      50 CONTINUE
C      RETURN
END
.....
C      PRINT OUT TYPICAL OUTPUT SUMMARY ON OUTPUT UNIT IPRINT
C      .....
SUBROUTINE RESULT(C, REGRES, IPRINT, IDIST, N, IDEP, NACT, NFIX,
1    XLOGL, NPAT, BETA, ISUB, ISTART, ICOV, COV, DET)
C      ARGUMENTS
C      LOGICAL REGRES
C      INTEGER IPRINT, IDIST, N, IDEP, NACT, NFIX, NPAT, ISUB(NPAT),
1    ISTART(NPAT), ICOV
C      DOUBLE PRECISION C, XLOGL, BETA(NPAT), COV(ICOV,NPAT), DET
C      LOCAL SCALARS
C      INTEGER IND
C      DOUBLE PRECISION TSTAT, ZERO
C      DATA ZERO / 0.0D0/

C      BASIC HEADINGS
C      MLE = C EQ ZERO
C      IF (MLE) WRITE (IPRINT,40)

```

```

C          FAILURE CODE - IFAULT = -1 - AN EXCEPTION WAS ABOUT TO
C          OCCUR WHILE PROCESSING THE I TH OBSERVATION
C          .....
C          SUBROUTINE DLOGBN(C, OBJECT, NACT, IACT, PAR, GRAD, HESS, N, IX,
1          X, IA, IFAULT)
C          ARGUMENTS
C          INTEGER NACT, IACT(NACT), N, IX, IA(N), IFAULT
C          REAL X(IX,N)
C          DOUBLE PRECISION C, OBJECT, PAR(NACT), GRAD(NACT), HESS(NACT,NACT)
C          LOCAL SCALARS
C          LOGICAL MLE
C          INTEGER IND
C          DOUBLE PRECISION EXTRA, F, S, G, FC, EITHER, DDENS, XJI, ZERO,
1          ONE, BIG
C          MACHINE-DEPENDENT CONSTANT - BIG ROUGHLY CHOSEN SO THAT
C          DEXP(X) WILL CAUSE EXCEPTION IF /X/ .GT. BIG
C          DATA ZERO, ONE, BIG /0.000, 1.000, 174.000/
C
C          IFAULT = 0
C          MLE = ONE + C .EQ. ONE
C          EXTRA = ONE / PAR(1)
C          OBJECT = ZERO
C          DO 10 J = 1, NACT
C            GRAD(J) = ZERO
C            DO 10 I = J, NACT
C              HESS(I,J) = ZERO
C            10 CONTINUE
C
C          MAIN LOOP OVER SAMPLE
C          DO 90 I = 1, N
C            FC = ZERO
C            DO 20 J = 1, NACT
C              IND = IACT(J)
C              FC = FC + PAR(J) * X(IND,I)
C            20 CONTINUE
C            IF (DABS(FC) .GT. BIG) GO TO 110
C            FC = DEXP(FC)
C            S = ONE + FC
C            F = FC / S
C            S = ONE / S
C            G = F * S
C            IF (IA(1) .EQ. 0) GO TO 30
C            EITHER = S
C            XJI = F
C            GO TO 40
C          30 EITHER = -F
C            XJI = S
C          40 IF (MLE) GO TO 50
C            FC = G ** C
C            GO TO 60
C          50 OBJECT = OBJECT + DLOG(XJI)
C            FC = ONE
C
C          LOOP TO INCREMENT GRADIENT AND HESSIAN UPPER TRIANGLE
C          60 EITHER = FC * EITHER
C            G = FC * G
C            FC = C * EITHER
C            DO 80 J = 1, NACT
C              IND = IACT(J)
C              XJI = DBLE(X(IND,I))

```

DLOG0005  
DLOG0006  
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DLOG0060  
DLOG0061  
DLOG0062  
DLOG0063  
DLOG0064

```

IF ( NOT MLE) WRITE (IPRINT,50) C
IF (IDIST EQ 1) WRITE (IPRINT,60)
IF (IDIST EQ 2) WRITE (IPRINT,70)
IF (IDIST EQ 3) WRITE (IPRINT,80)
WRITE (IPRINT,90) N, NPAR, NACT, NFIX
IF (REGRES) WRITE (IPRINT,100)
IF ( NOT REGRES) WRITE (IPRINT,110)

C
C
PARAMETER ESTIMATES
WRITE (IPRINT,120)
DO 20 I = 1, NPAR
  IND = ISUB(I)
  IF (ISTART(I) LT 2) GO TO 10
  WRITE (IPRINT,130) I, IND, BETA(I)
  GO TO 20
10 TSTAT = BETA(I) / COV(I,I)
  IF (IND EQ IDEP) WRITE (IPRINT,140) I, IND, BETA(I), COV(I,I),
1 TSTAT
  IF (IND NE IDEP) WRITE (IPRINT,150) I, IND, BETA(I), COV(I,I),
1 TSTAT
20 CONTINUE

C
C
ASYMPTOTIC QUANTITIES
IF (MLE) WRITE (IPRINT,160) XLOGL
IF (IDET LE ZERO) WRITE (IPRINT,170)
IF (IDET GT ZERO) WRITE (IPRINT,180) NACT, DET
WRITE (IPRINT,190)
DO 30 I = 1, NPAR
30 WRITE (IPRINT,200) (COV(I,J),J=1,I)

C
C
FORMAT ('MAXIMUM LIKELIHOOD BINARY ANALYSIS')
50 FORMAT ('SELF-CRITICAL BINARY ANALYSIS, C =', D11.3)
60 FORMAT ('LOGISTIC TOLERANCE DISTRIBUTION')
70 FORMAT ('GAUSSIAN TOLERANCE DISTRIBUTION')
80 FORMAT ('EXTREME VALUE TOLERANCE DISTRIBUTION')
90 FORMAT ('ANALYSIS OF', I6, ' CASES AND', I4, ' VARIABLES',/,'O',
1 14, ' PARAMETERS WERE ESTIMATED AND', I4,
2 ' WERE HELD CONSTANT', 2(/,'O'))
100 FORMAT ('PARAMETERS ARE EXPRESSED IN REGRESSION FORM',/,'O')
110 FORMAT ('PARAMETERS ARE EXPRESSED IN LOCATION-SCALE FORM',/,'O')
120 FORMAT (' I VAR', 8X, 'BETA(I)', 6X, 'STD ERR', 9X,
1 ' T STAT', 5X, 'STATUS')
130 FORMAT (2I5, D15.7, 35X, 'FIXED LOCATION')
140 FORMAT (2I5, 3D15.7, 5X, 'ESTIMATED SCALE')
150 FORMAT (2I5, 3D15.7, 5X, 'ESTIMATED LOCATION')
160 FORMAT (2(/,'O'), /, ' THE LOG LIKELIHOOD IS', D16.8)
170 FORMAT (2(/,'O'), /, ' THE ESTIMATED ASYMPTOTIC COVARIANCE',
1 ' MATRIX IS NOT POSITIVE DEFINITE')
180 FORMAT (2(/,'O'), /, ' THE', I4, ' ROOT OF THE DETERMINANT',
1 ' OF THE ESTIMATED',/,' ASYMPTOTIC COVARIANCE MATRIX IS',
2 D16.8)
190 FORMAT (/,' ESTIMATED ASYMPTOTIC STD ERRORS (ON DIAGONAL)',
1 ' , CORRELATIONS (BELOW DIAGONAL)',/,' ')
200 FORMAT (10D12.4)
RETURN
END

C.....DLG0001
C FIRST AND SECOND PARTIAL DERIVATIVES FOR LOGISTIC BINARY,
C REGRESSION PARAMETERIZATION PARTIALS ARE SCALED BY
C SAMPLE SIZE
C.....DLG0002
C.....DLG0003
C.....DLG0004

```

```

DDENS = XJI * EITHER
GRAD(J) = GRAD(J) + DDENS
DDENS = XJI * (S - F)
IF (J.EQ.1) DDENS = DDENS + EXTRA
DDENS = DDENS + FC - G * XJI
DO 70 K = J, NACT
  IND = IACT(K)
  HESS(K,J) = HESS(K,J) + X(IND,I) * DDENS
70 CONTINUE
80 CONTINUE
90 CONTINUE

C
C
C SCALE GRADIENT AND HESSIAN BY SAMPLE SIZE
EXTRA = DBLE(FLOAT(N))
DO 100 J = 1, NACT
  GRAD(J) = GRAD(J) / EXTRA
  DO 100 I = J, NACT
    HESS(I,J) = HESS(I,J) / EXTRA
    HESS(J,I) = HESS(I,J)
100 CONTINUE
IFAUULT = 0
RETURN

C
C
C ERROR EXIT
110 IFAULT = -1
RETURN

C
C
C
C.....
C COMPUTE V(*,*) FACTOR FOR ASYMPTOTIC COVARIANCE MATRIX.
C LOGISTIC BINARY. CALLED ONLY WHEN C.NE. 0
C.....
C SUBROUTINE VLOGBN(C, NACT, IACT, PAR, V, N, IX, X, IA)
C ARGUMENTS
C INTEGER NACT, IACT(NACT), N, IX, IA(N)
C REAL X(IX,N)
C DOUBLE PRECISION C, PAR(NACT), V(NACT,NACT)
C LOCAL SCALARS
C INTEGER IND
C DOUBLE PRECISION FC, S, F, EITHER, TEMP, ZERO, ONE, BIG
C MACHINE-DEPENDENT CONSTANT - BIG ROUGHLY CHOSEN SO THAT
C DEXP(X) WILL CAUSE EXCEPTION IF /X/.GT. BIG
C DATA ZERO, ONE, BIG / 0.000, 1.000, 174.000/

C
C
C SET UPPER TRIANGLE OF V(*,*) TO 0
DO 10 J = 1, NACT
  DO 10 I = J, NACT
    V(I,J) = ZERO
10 CONTINUE

C
C
C MAIN LOOP OVER SAMPLE
DO 50 I = 1, N
  FC = ZERO
  DO 20 J = 1, NACT
    IND = IACT(J)
    FC = FC + PAR(J) * X(IND,I)
20 CONTINUE
IF (DABS(FC) .GE. BIG) GO TO 50
FC = DEXP(FC)
S = ONE + FC
F = FC / S

```

```

S = ONE / S
FC = (F*S) ** C
EITHER = S
IF (IA(1) .EQ. 0) EITHER = F
EITHER = (EITHER*FC) ** 2

C
C
C INCREMENT TERM OF V(*,*)
DO 40 J = 1, NACT
  IND = IACT(J)
  TEMP = EITHER * X(IND,1)
  DO 30 K = J, NACT
    IND = IACT(K)
    V(K,J) = V(K,J) + TEMP * X(IND,1)
  30 CONTINUE
  40 CONTINUE
  50 CONTINUE

C
C
C DIVIDE V(*,*) BY SAMPLE SIZE, FILL OUT
TEMP = DBLE(FLOAT(N))
DO 60 J = 1, NACT
  DO 60 I = J, NACT
    V(I,J) = V(I,J) / TEMP
    V(J,1) = V(I,J)
  60 CONTINUE

C
C
C RETURN
END

C .....
C FIRST AND SECOND PARTIAL DERIVATIVES FOR BINARY GAUSSIAN.
C REGRESSION PARAMETERIZATION PARTIALS ARE SCALED BY
C SAMPLE SIZE
C FAILURE CODE - IFAULT = -1 - AN EXCEPTION WAS ABOUT TO
C OCCUR WHILE PROCESSING THE I TH OBSERVATION
C .....
SUBROUTINE DGAUBN(C, OBJECT, NACT, IACT, PAR, GRAD, HESS, N, IX,
1 X, IA, IFAULT)
C
C ARGUMENTS
C INTEGER NACT, IACT(NACT), N, IX, IA(N), IFAULT
C REAL X(IX,N)
C DOUBLE PRECISION C, OBJECT, PAR(NACT), GRAD(NACT), HESS(NACT,NACT)
C LOCAL SCALARS
C LOGICAL MLE
C INTEGER IND
C DOUBLE PRECISION EXTRA, DOT, DOTMIN, FC, CBY2, RATIO, EITHER,
1 HTERM, XJ1, DDENS, ZERO, ONE, TWO, BIG
C
C FUNCTIONS CALLED
C DOUBLE PRECISION ALNORM, RMILLS
C
C MACHINE-DEPENDENT CONSTANT - BIG ROUGHLY CHOSEN SO THAT
C DEXP(X) WILL CAUSE EXCEPTION IF /X/ .GT. BIG
C DATA ZERO, ONE, TWO, BIG / 0.000, 1.000, 2.000, 174.000/

C
C IFAULT = 0
C MLE = ONE + C .EQ. ONE
C EXTRA = ONE / PAR(1)
C CBY2 = C / TWO
C OBJECT = ZERO
C DO 10 J = 1, NACT
  GRAD(J) = ZERO
  DO 10 I = J, NACT
    HESS(I,J) = ZERO

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VLOG0033  
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DGAU0032  
DGAU0033



```

10 CONTINUE
C
C   MAIN LOOP OVER SAMPLE
DO 90 I = 1, N
  DOT = ZERO
  DO 20 J = 1, NACT
    IND = IACT(J)
    DOT = DOT + PAR(J) * X(IND,I)
  20 CONTINUE
  DOTMIN = -DOT
  IF (IA(1) .EQ. 0) GO TO 30
  RATIO = RMILLS(DOTMIN)
  EITHER = RATIO
  GO TO 40
  30 RATIO = RMILLS(DOT)
  DOT = DOTMIN
  EITHER = -RATIO
  IF (MLE) GO TO 50
  XJI = CBY2 * DOT * DOT
  IF (DABS(XJI) .GT. BIG) GO TO 110
  FC = DEXP(-XJI)
  GO TO 60
  50 FC = ONE
  XJI = ALNORM(DOT, .FALSE.)
  IF (XJI .EQ. ZERO) GO TO 110
  OBJECT = OBJECT + DLOG(XJI)
C
C   INCREMENT GRADIENT AND HESSIAN
  EITHER = FC * EITHER
  HTERM = -FC * RATIO * (DOT + RATIO)
  FC = C * EITHER
  DO 80 J = 1, NACT
    IND = IACT(J)
    XJI = DBLE(X(IND,I))
    GRAD(J) = GRAD(J) + XJI * EITHER
    DDENS = XJI * DOTMIN
    IF (J .EQ. 1) DDENS = DDENS + EXTRA
    DDENS = DDENS * FC + XJI * HTERM
    DO 70 K = J, NACT
      IND = IACT(K)
      HESS(K,J) = HESS(K,J) + X(IND,I) * DDENS
  70 CONTINUE
  80 CONTINUE
  90 CONTINUE
C
C   SCALE GRADIENT AND HESSIAN BY SAMPLE SIZE
  EXTRA = DBLE(FLOAT(N))
  DO 100 J = 1, NACT
    GRAD(J) = GRAD(J) / EXTRA
    DO 100 I = J, NACT
      HESS(I,J) = HESS(I,J) / EXTRA
      HESS(J,I) = HESS(I,J)
  100 CONTINUE
  IF ALT = 0
    RETURN
  C
  C   ERROR EXIT
  110 IF ALT = -1
    C
    RETURN

```

DGAU0034  
 DGAU0035  
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 DGAU0093

```

C.....
C      COMPUTE V(*.*) FACTOR FOR ASYMPTOTIC COVARIANCE MATRIX,
C      GAUSSIAN BINARY. CALLED ONLY WHEN C.NE.O
C.....
C      SUBROUTINE VGAUBN(C, NACT, IACT, PAR, V, N, IX, X, IA)
C.....
C      ARGUMENTS
C      INTEGER NACT, IACT(NACT), N, IX, IA(N)
C      REAL X(IX,N)
C      DOUBLE PRECISION C, PAR(NACT), V(NACT,NACT)
C      LOCAL SCALARS
C.....
C      INTEGER IND
C      DOUBLE PRECISION F2C, TEMP, EITHER, ZERO, BIG
C      FUNCTION CALLED
C.....
C      DOUBLE PRECISION RMILLS
C      MACHINE-DEPENDENT CONSTANT - BIG ROUGHLY CHOSEN SO THAT
C      DEXP(X) WILL CAUSE EXCEPTION IF /X/ GT. BIG
C      DATA ZERO, BIG /0.000, 174.000/
C.....
C      SET UPPER TRIANGLE OF V(*.*) TO ZERO
C      DO 10 J = 1, NACT
C        DO 10 I = J, NACT
C          10 V(I,J) = ZERO
C.....
C      MAIN LOOP OVER SAMPLE
C      DO 70 I = 1, N
C        F2C = ZERO
C        DO 20 J = 1, NACT
C          IND = IACT(J)
C          F2C = F2C + PAR(J) * X(IND,I)
C        20 CONTINUE
C        TEMP = C * F2C * F2C
C        IF (DABS(TEMP) GE. BIG) GO TO 70
C        IF (IA(I) EQ. 0) GO TO 30
C        EITHER = RMILLS(-F2C)
C        GO TO 40
C      30 EITHER = RMILLS(F2C)
C      40 F2C = DEXP(-TEMP) * EITHER * EITHER
C        DO 60 J = 1, NACT
C          IND = IACT(J)
C          TEMP = F2C * X(IND,I)
C          DO 50 K = J, NACT
C            IND = IACT(K)
C            V(K,I) = V(K,J) + TEMP * X(IND,I)
C          50 CONTINUE
C        60 CONTINUE
C        70 CONTINUE
C.....
C      DIVIDE V(*.*) BY SAMPLE SIZE, FILL OUT
C      TEMP = DBLE(FLOAT(N))
C      DO 80 J = 1, NACT
C        DO 80 I = J, NACT
C          V(I,J) = V(I,J) / TEMP
C          V(J,I) = V(I,J)
C        80 CONTINUE
C.....
C      RETURN
C      END
C.....
C      FIRST AND SECOND PARTIAL DERIVATIVES FOR BINARY EXTREME
C.....
DGAU0009
VGAU0001
VGAU0002
VGAU0003
VGAU0004
VGAU0005
VGAU0006
VGAU0007
VGAU0008
VGAU0009
VGAU0010
VGAU0011
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VGAU0052
VGAU0053
VGAU0054
VGAU0055
VGAU0056
VGAU0057
DEXV0001
DEXV0002

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```

C VALUE, REGRESSION PARAMETERIZATION. PARTIALS ARE SCALED DEXV0003
C BY SAMPLE SIZE. DEXV0004
C FAILURE CODE - IFAULT = -1 - AN EXCEPTION WOULD HAVE DEXV0005
C OCCURRED WHEN PROCESSING THE I TH OBSERVATION DEXV0006
C..... DEXV0007
C SUBROUTINE DEXVBN(C, OBJECT, NACT, IACT, PAR, GRAD, HESS, N, IX, DEXV0008
C X, IA, IFAULT) DEXV0009
C ARGUMENTS DEXV0010
C INTEGER NACT, IACT(NACT), N, IX, IA(N), IFAULT DEXV0011
C REAL X(IX,N) DEXV0012
C DOUBLE PRECISION C, OBJECT, PAR(NACT), GRAD(NACT), HESS(NACT,NACT) DEXV0013
C LOCAL SCALARS DEXV0014
C LOGICAL MLE DEXV0015
C INTEGER IND DEXV0016
C DOUBLE PRECISION EXTRA, DOT, EXPON, DENFAC, EXPEXP, FC, EITHER, DEXV0017
C HTERM, ZERO, ONE, BIG, BIG1 DEXV0018
C MACHINE-DEPENDENT CONSTANTS - BIG ROUGHLY CHOSEN SO THAT DEXV0019
C DEXP(X) WILL CAUSE EXCEPTION IF /X/ .GT. BIG, BIG1 SUCH DEXV0020
C THAT 1.000 / DEXP(DEXP(X)) NE. 1.000 IF X .GT. BIG1 DEXV0021
C DATA ZERO, ONE, BIG, BIG1 / 0.000, 1.000, 174.000, -36.200/ DEXV0022
C IFAULT = 0 DEXV0023
C MLE = ONE + C .EQ. ONE DEXV0024
C EXTRA = ONE / PAR(1) DEXV0025
C OBJECT = ZERO DEXV0026
C DO 10 J = 1, NACT DEXV0027
C GRAD(J) = ZERO DEXV0028
C DO 10 I = J, NACT DEXV0029
C HESS(I,J) = ZERO DEXV0030
C 10 CONTINUE DEXV0031
C DEXV0032
C DEXV0033
C DEXV0034
C DEXV0035
C DEXV0036
C DEXV0037
C DEXV0038
C DEXV0039
C DEXV0040
C DEXV0041
C DEXV0042
C DEXV0043
C DEXV0044
C DEXV0045
C DEXV0046
C DEXV0047
C DEXV0048
C DEXV0049
C DEXV0050
C DEXV0051
C DEXV0052
C DEXV0053
C DEXV0054
C DEXV0055
C DEXV0056
C DEXV0057
C DEXV0058
C DEXV0059
C DEXV0060
C DEXV0061
C DEXV0062

C VALUE, REGRESSION PARAMETERIZATION. PARTIALS ARE SCALED
C BY SAMPLE SIZE.
C FAILURE CODE - IFAULT = -1 - AN EXCEPTION WOULD HAVE
C OCCURRED WHEN PROCESSING THE I TH OBSERVATION
C.....
C SUBROUTINE DEXVBN(C, OBJECT, NACT, IACT, PAR, GRAD, HESS, N, IX,
C X, IA, IFAULT)
C ARGUMENTS
C INTEGER NACT, IACT(NACT), N, IX, IA(N), IFAULT
C REAL X(IX,N)
C DOUBLE PRECISION C, OBJECT, PAR(NACT), GRAD(NACT), HESS(NACT,NACT)
C LOCAL SCALARS
C LOGICAL MLE
C INTEGER IND
C DOUBLE PRECISION EXTRA, DOT, EXPON, DENFAC, EXPEXP, FC, EITHER,
C HTERM, ZERO, ONE, BIG, BIG1
C MACHINE-DEPENDENT CONSTANTS - BIG ROUGHLY CHOSEN SO THAT
C DEXP(X) WILL CAUSE EXCEPTION IF /X/ .GT. BIG, BIG1 SUCH
C THAT 1.000 / DEXP(DEXP(X)) NE. 1.000 IF X .GT. BIG1
C DATA ZERO, ONE, BIG, BIG1 / 0.000, 1.000, 174.000, -36.200/
C IFAULT = 0
C MLE = ONE + C .EQ. ONE
C EXTRA = ONE / PAR(1)
C OBJECT = ZERO
C DO 10 J = 1, NACT
C GRAD(J) = ZERO
C DO 10 I = J, NACT
C HESS(I,J) = ZERO
C 10 CONTINUE
C
C MAIN LOOP OVER SAMPLE
C DO 90 I = 1, N
C DOT = ZERO
C DO 20 J = 1, NACT
C IND = IACT(J)
C DOT = DOT + PAR(J) * X(IND,I)
C 20 CONTINUE
C IF (DABS(DOT) GE BIG) GO TO 110
C EXPON = DEXP(DOT)
C DENFAC = ONE - EXPON
C IF (IA(I) EQ. 0) GO TO 40
C
C FAILURE TERM, IA(I) = 1
C IF (EXPON GE BIG OR DOT LE BIG1) GO TO 110
C EXPEXP = DEXP(EXPON)
C DOT = ONE - ONE / EXPEXP
C FC = EXPON / EXPEXP
C EITHER = FC / DOT
C HTERM = EITHER * (DENFAC - EITHER)
C IF (MLE) GO TO 30
C FC = FC ** C
C GO TO 60
C 30 FC = ONE
C OBJECT = OBJECT + DLOG(DOT)
C GO TO 60
C
C NON FAILURE TERM, IA(I) = 0
C EITHER = EXPON
C HTERM = EITHER

```

[illegible]

```

C
C
      DO 10 I = J, NACT
      IO V(I,J) = ZERO
C
C      MAIN LOOP OVER SAMPLE
      DO 70 I = 1, N
      DOT = ZERO
      DO 20 J = 1, NACT
      IND = 1ACT(J)
      DOT = DOT + PAR(J) * X(IND,I)
      CONTINUE
20   IF (DABS(DOT) GE BIG) GO TO 70
      EXPON = DEXP(DOT)
      IF (IA(I) EQ O) GO TO 30
      IF (DOT LT BIGI) GO TO 70
      FC = DEXP(EXPON)
      DOT = ONE - ONE / FC
      FC = EXPON / FC
      EXPON = FC / DOT
      FC = EXPON * FC ** C
      GO TO 40
30   FC = C * (DOT - EXPON)
      IF (DABS(FC) GT BIG) GO TO 70
      FC = EXPON * DEXP(FC)
C
C      SUMMANDS FOR V(*.*)
      FC = FC * FC
      DO 60 J = 1, NACT
      IND = 1ACT(J)
      DOT = FC * X(IND,I)
      DO 50 K = J, NACT
      IND = 1ACT(K)
      V(K,J) = V(K,J) + DOT * X(IND,I)
50   CONTINUE
60   CONTINUE
70   CONTINUE
C
C      DIVIDE V(*.*) BY SAMPLE SIZE, FILL OUT
      DOT = DBLE(FLOAT(N))
      DO 80 J = 1, NACT
      DO 80 I = J, NACT
      V(I,J) = V(I,J) / DOT
      V(J,I) = V(I,J)
80   CONTINUE
C
C      RETURN
      END
C.....
C      DOUBLE PRECISION MATRIX MULTIPLICATION
C      X(N1 BY N3) = Y(N1 BY N2) * Z(N2 BY N3)
C      THREE OPTIONS -
C      IFLAG = 0 - X, Y, AND Z ARE DISTINCT
C      IFLAG LT 0 - X, Y OVERLAP, CORNER OF Y OVERWRITTEN
C      IFLAG GT 0 - X, Z OVERLAP, CORNER OF Z OVERWRITTEN
C.....
C      SUBROUTINE DMXMLT(X, IX, N1, Y, IY, N2, Z, IZ, N3, WORK, LWORK,
1         IFLAG, IFAULT)
C      ARGUMENTS
C      INTEGER IX, N1, IY, N2, IZ, N3, LWORK, IFLAG, IFAULT
C      DOUBLE PRECISION X(IX,N3), Y(IY,N2), Z(IZ,N3), WORK(LWORK)
C      LOCAL SCALARS

```

```

C      DOUBLE PRECISION TEMP, ZERO
C      DATA ZERO /0.000/
C
C      ERROR EXITS
C      IFAULT = 1
C      IF (MINO(N1,N2,N3) .LT. 1 .OR. MINO(IX,IY) .LT. N1 .OR. IZ .LT.
C        1 N2) RETURN
C      IFAULT = 2
C      IF ((IFLAG .LT. 0 .AND. (N3 .GT. N2 .OR. LWORK .LT. N3))
C        1 .RETURN
C      IFAULT = 3
C      IF ((IFLAG .GT. 0 .AND. (N1 .GT. IZ .OR. LWORK .LT. N1))
C        1 .RETURN
C      IFAULT = 0
C      IF ((IFLAG .NE. 0) GO TO 30
C
C      STRAIGHTFORWARD, NO OVERWRITING
C      DO 20 I = 1, N1
C        DO 20 J = 1, N3
C          TEMP = ZERO
C          DO 10 K = 1, N2
C            TEMP = TEMP + Y(I,K) * Z(K,J)
C          X(I,J) = TEMP
C        10 CONTINUE
C      20 CONTINUE
C      RETURN
C
C      CORNER OF MATRIX Z IS OVERWRITTEN
C      30 IF ((IFLAG .LT. 0) GO TO 80
C        DO 70 J = 1, N3
C          DO 50 I = 1, N1
C            TEMP = ZERO
C            DO 40 K = 1, N2
C              TEMP = TEMP + Y(I,K) * Z(K,J)
C            WORK(I) = TEMP
C          50 CONTINUE
C          DO 60 I = 1, N1
C            X(I,J) = WORK(I)
C          60 CONTINUE
C        70 CONTINUE
C      RETURN
C
C      CORNER OF MATRIX Y IS OVERWRITTEN
C      80 DO 120 I = 1, N1
C        DO 100 J = 1, N3
C          TEMP = ZERO
C          DO 90 K = 1, N2
C            TEMP = TEMP + Y(I,K) * Z(K,J)
C          WORK(J) = TEMP
C        100 CONTINUE
C        DO 110 J = 1, N3
C          X(I,J) = WORK(J)
C        120 CONTINUE
C      RETURN
C      END
C.....
C      DOUBLE PRECISION VERSION OF ALGORITHM A5 66, APPLIED
C      STATISTICS (1973), VOL 22, NO 3
C      EVALUATES THE TAIL AREA OF THE STANDARD NORMAL CURVE
C      FROM X TO INFINITY IF UPPER IS TRUE OR FROM
C      MINUS INFINITY TO X IF UPPER IS FALSE

```

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C.....ALND0007
C DOUBLE PRECISION FUNCTION ALNORM(X,UPPER)
C ARGUMENTS
C LOGICAL UPPER
C DOUBLE PRECISION X
C LOCAL SCALARS
C LOGICAL UP
C DOUBLE PRECISION LTONE, UTZERO, ZERO, HALF, ONE, CON, Z, Y
C MACHINE-DEPENDENT CONSTANTS - LTONE = (N + 9) / 3, WHERE
C N IS NO. OF DECIMAL DIGITS IN DOUBLE PRECISION NUMBER,
C UTZERO SUCH THAT DEXP(-X*X/2.000) WILL CAUSE AN EXCEPTION
C IF X .GT. UTZERO
C CONSTANTS IN EXPRESSIONS ARE AS IN AS 66
C DATA LTONE, UTZERO, ZERO, HALF, ONE, CON /8.000, 18.6500, 0.000,
C 1 0.500, 1.000, 1.2800/
C
C UP = UPPER
C Z = X
C IF (Z .GE. ZERO) GO TO 10
C UP = NOT UP
C Z = -Z
C 10 IF (Z .LE. LTONE .OR. UP .AND. Z .LE. UTZERO) GO TO 20
C ALNORM = ZERO
C GO TO 40
C 20 Y = HALF * Z * Z
C IF (Z .GT. CON) GO TO 30
C
C ALNORM = HALF - Z * (0.3989422804400 - 0.39990343850400*Y/(Y + 5.
C 17588548045800 - 29.821355780800/(Y + 2.62433312167900+48.
C 2695953069200/(Y + 5.9288572443800))))
C GO TO 40
C
C 30 ALNORM = 0.39894228038500 * DEXP(-Y) / (Z - 3.80520 - 8 + 1.
C 10000061530200/(Z + 3.980647940 - 4 + 1.9861538136400/(Z - 0.
C 215167911663500+5.2933032492600/(Z + 4.838591280800 - 15.
C 3150897245100/(Z + 0.74238092402700+30.78993303400/(Z + 3.
C 49901941701100))))))
C
C 40 IF (NOT UP) ALNORM = ONE - ALNORM
C
C RETURN
C END
C.....RMIL0001
C RECIPROCAL OF MILLS RATIO, Z(X) / Q(X), X A STANDARD NORMAL
C VARIATE, BASED ON FUNCTION ALNORM(), ALGORITHM AS 66.
C.....RMIL0003
C DOUBLE PRECISION FUNCTION RMILLS(X)
C ARGUMENTS
C DOUBLE PRECISION X
C LOCAL SCALARS
C LOGICAL UP
C DOUBLE PRECISION Z, Y, LTONE, UTZERO, ZERO, HALF, ONE, CON, FPI1,
C 1 FPI2
C MACHINE-DEPENDENT CONSTANTS - LTONE = (N + 9) / 3, WHERE
C N IS NO. OF DECIMAL DIGITS IN DOUBLE PRECISION NUMBER,
C UTZERO SUCH THAT DEXP(-X*X/2.000) WILL CAUSE AN EXCEPTION
C IF X .LT. UTZERO
C FPI1 = SQRT(2/PI), FPI2 = 1/SQRT(2*PI)
C CONSTANTS IN EXPRESSIONS ARE AS IN AS 66
C DATA LTONE, UTZERO, ZERO, HALF, ONE, CON, FPI1, FPI2 /8.000,

```

```

1 -18 6500. 0 000. 0 500. 1 000. 1 2800. 7 9788456080286536D-1. RM1L0019
2 3 9894228040143268D-1/ RM1L0020
C RM1L0021
C RM1L0022
C RM1L0023
C RM1L0024
C RM1L0025
C RM1L0026
C RM1L0027
C RM1L0028
C RM1L0029
C RM1L0030
C RM1L0031
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C RM1L0060
C RM1L0061
C RM1L0062
C RM1L0063
C RM1L0064
C RM1L0065
C RM1L0066
C RM1L0067
C RM1L0068

1 -18 6500. 0 000. 0 500. 1 000. 1 2800. 7 9788456080286536D-1. RM1L0019
2 3 9894228040143268D-1/ RM1L0020
C RM1L0021
C RM1L0022
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C RM1L0060
C RM1L0061
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C RM1L0063
C RM1L0064
C RM1L0065
C RM1L0066
C RM1L0067
C RM1L0068

TRIVIAL CASES
IF (X NE ZERO) GO TO 10
RMILLS = FPI1
RETURN
10 IF (X GT UTZERO) GO TO 20
RMILLS = ZERO
RETURN

USUAL SITUATION
20 UP = TRUE.
Z = X
IF (Z GE ZERO) GO TO 30
UP = FALSE.
Z = -Z
30 Y = HALF * Z * Z
IF (Z GT CON) GO TO 50

CENTRAL PORTION - O LT ABS(X) LT 1.28
RMILLS = Z * (O 398942280444DO-O 399903438504DO*(Y + 5
175885480458DO-29 8213557808DO/(Y + 2 62433121679DO+48
26959930692DO/(Y + 5 92885724438DO))))
Y = FPI2 * DEXP(-Y)
IF (UP) GO TO 40
RMILLS = Y / (HALF + RMILLS)
RETURN
40 RMILLS = Y / (HALF - RMILLS)
RETURN

OUTER PORTION - ABS(X) GE 1.28
50 IF (UP OR Z LE LTONE) GO TO 60

SPECIAL CASE - LOWER TAIL AND Q ESSENTIALLY 1
RMILLS = FPI2 * DEXP(-Y)
RETURN

USUAL SITUATION
60 RMILLS = (Z - 3 8052D 8 + 1 00000615302DO/(Z + 3 98064794D-4 + 1
198615381364DO/(Z - O 151679116635DO*5 29330324926DO/(Z + 4
28385912808DO 15 1508972451DO/(Z + O 742380924027DO+30 789933034DO/
3(Z + 3 99019417011DO))))))

IF (UP) RETURN
Y = FPI2 * DEXP(-Y)
RMILLS = Y / (ONE - Y/RMILLS)
RETURN
END

```



```

C.....MAIN0001
C   SAMPLE MAIN PROGRAM AND INPUT SUBROUTINE FOR SELF-CRITICAL
C   BINARY ESTIMATION. IT IS RECOMMENDED THAT BINARY() AND
C   ITS AUXILIARY PROCEDURES BE COMPILED AND PLACED IN AN
C   OBJECT CODE LIBRARY, SO THE USER CAN CALL BINARY() FROM
C   ARBITRARY PROGRAMS.
C.....MAIN0002
C   INTEGER N, IX, IA(750), NPAR, ISUB(10), ISTART(10), IDEP, IDIST,
C   1   MAXIT, IPRINT, IFLAG, ICOV, LMEM, MEMORY(490), IFAULT, NC,
C   2   NMAX, NCMAX, IREAD
C   REAL X(10,750)
C   DOUBLE PRECISION C(5), RELTOL, ABSTOL, BETA(10), XLOGL, COV(10,10)
C.....MAIN0003
C   DIMENSION SPECIFICATIONS FOR ARRAYS.
C   THIS MAIN PROGRAM CAN HANDLE UP TO 750 OBSERVATIONS,
C   WITH UP TO 10 PARAMETERS AND 5 DIFFERENT VALUES OF
C   C FOR ESTIMATION.
C   DATA IX, ICOV, LMEM, NMAX, NCMAX /10, 10, 490, 750, 5/
C.....MAIN0004
C   BASIC INFORMATION TO PASS TO BINARY() - IPRINT STANDS
C   FOR LOGICAL OUTPUT UNIT 6
C   DATA MAXIT, RELTOL, ABSTOL, IPRINT, IFLAG /15, 1.0D-7, 1.0D-7, 6,
C   1   0/
C   IREAD IS LOGICAL INPUT UNIT 5
C   DATA IREAD /5/
C.....MAIN0005
C   SPECIFY EVERYTHING BY HAND IN SUBROUTINE INPUT - ARRAY
C   MEMORY(*) PASSED AS WORKSPACE
C   CALL INPUT(IX, N, NMAX, X, IA, NPAR, BETA, ISUB, ISTART, NC,
C   1   NCMAX, C, IDEP, IDIST, MEMORY, IREAD, IPRINT, IFAULT)
C   IF (IFAUULT EQ 0) GO TO 10
C   IF (IFAUULT EQ 1) WRITE (IPRINT,50) N, NMAX
C   IF (IFAUULT EQ 2) WRITE (IPRINT,60) NPAR, IX
C   IF (IFAUULT EQ 3) WRITE (IPRINT,70) NC, NCMAX
C   WRITE (IPRINT,80)
C   STOP
C.....MAIN0006
C   LOOP OVER THE VALUES OF C REQUESTED FOR ESTIMATION
C   10 DO 40 I = 1, NC
C       CALL BINARY(N, IX, X, IA, NPAR, ISUB, ISTART, IDEP, IDIST, C(I),
C   1   RELTOL, ABSTOL, MAXIT, IPRINT, IFLAG, BETA, XLOGL, ICOV,
C   2   COV, LMEM, MEMORY, IFAULT)
C       IF (IFAUULT EQ 0) GO TO 20
C       WRITE (IPRINT,90) IFAULT
C       STOP
C   20 IF (I .GT. 1) GO TO 40
C.....MAIN0007
C   AFTER THE FIRST ESTIMATION, SET ISTART(*) TO 1, SO
C   LATEST ESTIMATES CAN BE USED AS STARTING VALUES
C   DO 30 J = 1, NPAR
C   30 ISTART(J) = 1
C   40 CONTINUE
C.....MAIN0008
C   50 FORMAT ('OERROR - SAMPLE SIZE OF ', I10, ' EXCEEDS DIMENSION OF ',
C   1   I10)
C   60 FORMAT ('OERROR - ', I3, ' PARAMETERS EXCEEDS DIMENSION OF ', I3)
C   70 FORMAT ('OERROR - ', I3, ' C VALUES EXCEEDS DIMENSION OF ', I3)
C   80 FORMAT ('ORECOMPILE MAIN PROGRAM WITH NEW DIMENSIONS')

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```

90 FORMAT (' ESTIMATION PROCEDURE HAS FAILED - IFAULT = ', I6)
STOP
END
C.....
C      CRUDE INPUT ROUTINE  SETS ESTIMATION CONTROL PARAMETERS
C      EXPLICITLY, INSTEAD OF READING THEM IN
C.....
C      SUBROUTINE INPUT(IX, N, NMAX, X, IA, NPAR, BETA, ISUB, ISTART, NC,
1      INPU0005      INPU0006      INPU0007      INPU0008      INPU0009      INPU0010      INPU0011      INPU0012      INPU0013      INPU0014      INPU0015      INPU0016      INPU0017      INPU0018      INPU0019      INPU0020      INPU0021      INPU0022      INPU0023      INPU0024      INPU0025      INPU0026      INPU0027      INPU0028      INPU0029      INPU0030      INPU0031      INPU0032      INPU0033      INPU0034      INPU0035      INPU0036      INPU0037      INPU0038      INPU0039      INPU0040      INPU0041      INPU0042      INPU0043      INPU0044      INPU0045      INPU0046      INPU0047      INPU0048      INPU0049      INPU0050      INPU0051      INPU0052      INPU0053      INPU0054      INPU0055      INPU0056      INPU0057
      NCMAX, C, IDEP, IDIST, WORK, IREAD, IPRINT, IFAULT)
C      ARGUMENTS
C      INTEGER IX, N, NMAX, IA(1), NPAR, ISUB(1), ISTART(1), NC, NCMAX,
1      IDEP, IDIST, IPRINT, IFAULT
C      REAL X(IX,1), WORK(1)
C      DOUBLE PRECISION BETA(1), C(1)
C      LOCAL TYPE DECLARATION
C      LOGICAL FLAG
C
C      SET DETAILS OF PROBLEM SIZE
C      N = 680
C      NPAR = 5
C      NC = 4
C
C      CHECK FOR SIZE ERRORS
C      IFAULT = 1
C      IF (N GT NMAX) RETURN
C      IFAULT = 2
C      IF (NPAR GT IX) RETURN
C      IFAULT = 3
C      IF (NC GT NCMAX) RETURN
C      IFAULT = 0
C
C      MODELING DETAILS
C      DEPENDENT VARIABLE IN FIRST ROW
C      IDEP = 1
C      CONSTANT IN SECOND ROW
C      ICONST = 2
C      GAUSSIAN TOLERANCE DISTRIBUTION
C      IDIST = 2
C      VALUES OF C FOR SELF-CRITICAL
C      C(1) = 0.000
C      C(2) = 0.100
C      C(3) = 0.200
C      C(4) = 0.300
C
C      INITIALIZE AVERAGES TO 0
C      YBAR = 0.0
C      YSE = 0.0
C      DO 10 I = 1, NPAR
10      WORK(I) = 0.0
C
C      LOOP OVER SAMPLE TO READ IN DATA
C      FLAG = NPAR * LE 2
C      DO 40 I = 1, N
      READ (IREAD,20) VC, IA(I), ALPHA, S
20      FORMAT (F10.0, I2, 2F8.0)
C      EXPRESS S IN METERS, SCALE VC
C      VC = ALOG(VC/1000.0)
C      S = 0.3048 * EXP(S)
C

```

```

C      C      FILL THE DATA MATRIX - CONSTANT IN SECOND ROW
      X(IDEP,1) = VC
      X(ICONST,1) = 1.0
      X(3,1) = ALPHA
      X(4,1) = S
      X(5,1) = ALPHA * S

C      C      UPDATE MEAN AND STD ERROR OF DEPENDENT VARIABLE IN A
C      C      WAY WHICH DOESNT LOSE SIGNIFICANT FIGURES
      TEMP = VC - YBAR
      YBAR = YBAR + TEMP / FLOAT(1)
      YSE = YSE + TEMP * (VC - YBAR)

C      C      INCREMENT COVARIATE AVERAGES (3D ROW AND UP)
      IF (FLAG) GO TO 40
      DO 30 J = 3, NPAR
      30  WORK(J) = WORK(J) + X(J,1)
C      40 CONTINUE

C      C      SET ISUB(*) AND ISTART(*), PROVIDING STARTING VALUES
C      C      FOR INTERCEPT AND SCALE
      DO 50 I = 1, NPAR
      50  ISUB(I) = 1
      50  ISTART(I) = 0
      50  CONTINUE
      TEMP = FLOAT(N)
      YSE = SORT(YSE/TEMP)
      ISTART(IDEP) = 1
      BETA(IDEP) = DBLE(YSE)
      ISTART(ICONST) = 1
      BETA(ICONST) = DBLE(YBAR)

C      C      WRITE OUT INFORMATION ON STD. ERROR AND AVERAGES
      WRITE (IPRINT,60) YBAR, YSE
      60  FORMAT ('DEPENDENT (STRESS) VARIABLE'/'MEAN =', E14.6,
      1  'STANDARD DEVIATION =', E14.6)
      IF (FLAG) RETURN
      WRITE (IPRINT,70)
      70  FORMAT ('COVARIATES HAVE BEEN CENTERED BY THEIR MEANS'/'
      1  'OVARIBLE', 10X, 'MEAN')
      DO 90 I = 3, NPAR
      90  WORK(I) = WORK(I) / TEMP
      WRITE (IPRINT,80) ISUB(1), WORK(1)
      80  FORMAT ('O', 18, E14.6)
      90  CONTINUE

C      C      SUBTRACT MEANS FROM COVARIATES
      DO 110 I = 1, N
      110  X(J,I) = X(J,I) - WORK(J)
      100  X(J,I) = X(J,I) - WORK(J)
      110 CONTINUE
C      RETURN
      END

```

INPU0058  
 INPU0059  
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AD-A178545

## REPORT DOCUMENTATION PAGE

READ INSTRUCTIONS  
BEFORE COMPLETING FORM

1. REPORT NUMBER ARO-3	2. GOVT ACCESSION NO. N/A	3. RECIPIENT'S CATALOG NUMBER N/A
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